EXPT NO: 1 A python program to implement univariate regression

DATE:22/08/2024 bivariate regression and multivariate regression.

AIM:

To write a python program to implement univariate regression, bivariate regression and multivariate regression.

PROCEDURE:

Implementing univariate, bivariate, and multivariate regression using the Iris dataset involve the following steps:

Step 1: Import Necessary Libraries

First, import the libraries that are essential for data manipulation, visualization, and model building.

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean squared error, r2 score
```

Step 2: Load the Iris Dataset

The Iris dataset can be loaded and display the first few rows of the dataset .

```
# Load the Iris dataset
iris = sns.load_dataset('iris')
# Display the first few rows of the dataset
print(iris.head())
```

	Sebat Mincu	petal_length	petal_width	species
5.1	3.5	1.4	0.2	setosa
4.9	3.0	1.4	0.2	setosa
4.7	3.2	1.3	0.2	setosa
4.6	3.1	1.5	0.2	setosa
5.0	3.6	1.4	0.2	setosa
	4.9 4.7 4.6	4.9 3.0 4.7 3.2 4.6 3.1	4.9 3.0 1.4 4.7 3.2 1.3 4.6 3.1 1.5	4.9 3.0 1.4 0.2 4.7 3.2 1.3 0.2 4.6 3.1 1.5 0.2

Step 3: Data Preprocessing

Ensure the data is clean and ready for modeling. Since the Iris dataset is clean, minimal preprocessing is needed.

```
# Check for missing values
print(iris.isnull().sum())

# Display the basic statistical details
print(iris.describe())
```

OUTPUT:

```
→ sepal length
   sepal width
   petal length
   petal width
                 0
   species
                 0
   dtype: int64
         sepal_length sepal_width petal_length petal_width
   count
           150.000000 150.000000
                                  150.000000 150.000000
   mean
            5.843333
                       3.057333
                                   3.758000
                                              1.199333
            0.828066
                                   1.765298
   std
                        0.435866
                                              0.762238
           4.300000
   min
                       2.000000
                                   1.000000
                                              0.100000
   25%
            5.100000
                       2.800000
                                   1.600000
                                              0.300000
            5.800000
                       3.000000
                                    4.350000
                                              1.300000
   50%
   75%
            6.400000
                       3.300000
                                   5.100000
                                              1.800000
            7,900000 4,400000 6,900000 2,500000
   max
```

Step 4: Univariate Regression

Univariate regression involves predicting one variable based on a single predictor.

4.1 : Select the Features

Choose one feature (e.g., sepal_length) and one target variable (e.g., sepal_width).

```
X_uni = iris[['sepal_length']]
y uni = iris['sepal width']
```

4.2 : Split the Data

Split the data into training and testing sets.

Fit the linear regression model on the training data.

```
X_uni_train, X_uni_test, y_uni_train, y_uni_test = train_test_split(X_uni,
y_uni,
test size=0.2, random state=42)
```

4.3: Train the model

```
uni_model = LinearRegression()
uni model.fit(X uni train, y uni train)
```



```
* LinearRegression 0 0 0 LinearRegression()
```

4.4 : Make Predictions

Use the model to make predictions on the test data.

```
y uni pred = uni model.predict(X uni test)
```

4.5 : Evaluate the Model

Evaluate the model performance using metrics like Mean Squared Error (MSE) and R-squared.

```
print(f'Univariate MSE: {mean_squared_error(y_uni_test, y_uni_pred)}')
print(f'Univariate R-squared: {r2_score(y_uni_test, y_uni_pred)}')
```

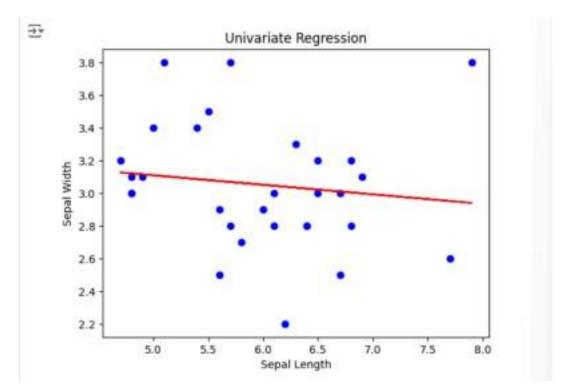
OUTPUT:

```
Univariate MSE: 0.13961895650579023
Univariate R-squared: 0.024098626473972984
```

4.6: Visualize the Results

Visualize the relationship between the predictor and the target variable.

```
plt.scatter(X_uni_test, y_uni_test, color='blue')
plt.plot(X_uni_test, y_uni_pred, color='red')
plt.xlabel('Sepal Length')
plt.ylabel('Sepal Width')
plt.title('Univariate Regression')
plt.show()
```



Step 5 : Bivariate Regression

Bivariate regression involves predicting one variable based on two predictors.

5.1: Select the Features

Choose two features (e.g., sepal_length, petal_length) and one target variable (e.g., sepal_width).

```
X_bi = iris[['sepal_length', 'petal_length']]
y_bi = iris['sepal_width']
```

5.2 : Split the Data

Split the data into training and testing sets.

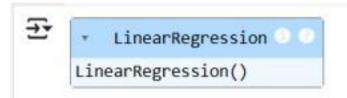
```
X_bi_train, X_bi_test, y_bi_train, y_bi_test = train_test_split(X_bi, y_bi,
test size=0.2, random state=42)
```

5.3: Train the Model

Fit the linear regression model on the training data.

```
bi_model = LinearRegression()
bi model.fit(X bi train, y bi train)
```

OUTPUT:



5.4: Make Predictions

Use the model to make predictions on the test data.

```
y_bi_pred = bi_model.predict(X_bi_test)
```

5.5 : Evaluate the Model

Evaluate the model performance using metrics like MSE and R-squared.

```
print(f'Bivariate MSE: {mean_squared_error(y_bi_test, y_bi_pred)}')
print(f'Bivariate R-squared: {r2_score(y_bi_test, y_bi_pred)}')
```

OUTPUT:

```
Bivariate MSE: 0.08308605032913309
Bivariate R-squared: 0.4192494152204116
```

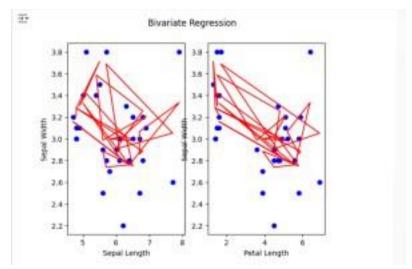
5.6: Visualize the Results

Since visualizing in 3D is challenging, we can plot the relationships between the target and each predictor separately.

```
# Sepal Length vs Sepal Width
plt.subplot(1, 2, 1)
plt.scatter(X_bi_test['sepal_length'], y_bi_test, color='blue')
plt.plot(X_bi_test['sepal_length'], y_bi_pred, color='red')
plt.xlabel('Sepal Length')
plt.ylabel('Sepal Width')
# Petal Length vs Sepal Width
plt.subplot(1, 2, 2)
```

```
plt.scatter(X_bi_test['petal_length'], y_bi_test, color='blue')
plt.plot(X_bi_test['petal_length'], y_bi_pred, color='red')
plt.xlabel('Petal Length')
plt.ylabel('Sepal Width')
plt.suptitle('Bivariate Regression')
plt.show()
```

OUTPUT:



Step 6: Multivariate Regression

Multivariate regression involves predicting one variable based on multiple predictors.

6.1: Select the Features

Choose multiple features (e.g., sepal_length, petal_length, petal_width) and one target variable (e.g., sepal_width).

```
X_multi = iris[['sepal_length', 'petal_length', 'petal_width']]
y_multi = iris['sepal_width']
```

6.2 : Split the Data

Split the data into training and testing sets.

```
X_multi_train, X_multi_test, y_multi_train, y_multi_test =
```

```
train_test_split(X_multi,

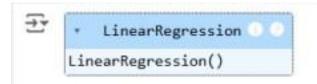
y_multi, test_size=0.2, random_state=42)
```

6.3: Train the Model

Fit the linear regression model on the training data.

```
multi_model = LinearRegression()
multi_model.fit(X_multi_train, y_multi_train)
```

OUTPUT:



6.4: Make Predictions

Use the model to make predictions on the test data.

```
y_multi_pred = multi_model.predict(X_multi_test)
```

6.5 : Evaluate the Model

Evaluate the model performance using metrics like MSE and R-squared.

```
print(f'Multivariate MSE: {mean_squared_error(y_multi_test, y_multi_pred)}')
print(f'Multivariate R-squared: {r2_score(y_multi_test, y_multi_pred)}')
```

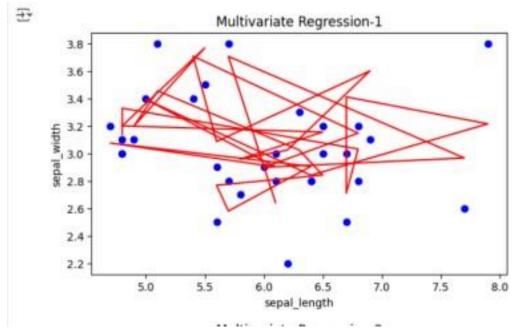
OUTPUT:

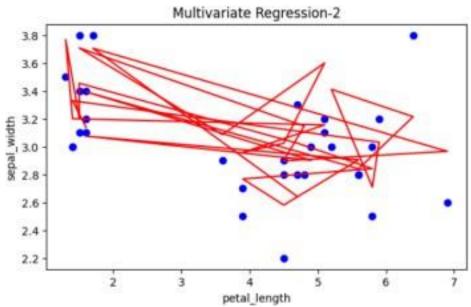
```
Multivariate MSE: 0.0868353771078583
Multivariate R-squared: 0.39304256448374897
```

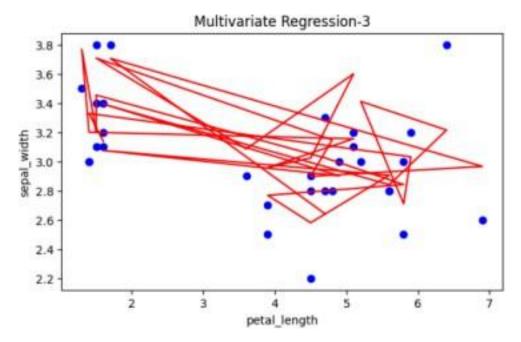
Step 7: Visualize the multivariate regression

```
plt.figure(figsize=(15,4))
plt.subplot(1, 2, 1)
plt.scatter(X_multi_test['sepal_length'], y_multi_test, color='blue')
plt.plot(X_multi_test['sepal_length'], y_multi_pred, color='red')
plt.xlabel('sepal_length')
plt.ylabel('sepal_width')
```

```
plt.title('Multivariate Regression-1')
plt.show()
plt.figure(figsize=(15,4))
plt.subplot(1, 2, 1)
plt.scatter(X_multi_test['petal_length'], y_multi_test, color='blue')
plt.plot(X_multi_test['petal_length'], y_multi_pred, color='red')
plt.xlabel('petal_length')
plt.ylabel('sepal_width')
plt.title('Multivariate Regression-2')
plt.show()
plt.figure(figsize=(15,4))
plt.subplot(1, 2, 2)
plt.scatter(X_multi_test['petal_length'], y_multi_test, color='blue')
plt.plot(X multi test['petal length'], y multi pred, color='red')
plt.xlabel('petal length')
plt.ylabel('sepal_width')
plt.title('Multivariate Regression-3')
plt.show()
```







Step 8: Interpret the Results

After implementing and evaluating the models, interpret the coefficients to understand the influence of each predictor on the target variable.

```
print('Univariate Coefficients:', uni_model.coef_)
print('Bivariate Coefficients:', bi_model.coef_)
print('Multivariate Coefficients:', multi model.coef )
```

```
Univariate Coefficients: [-0.05829418]

Bivariate Coefficients: [ 0.56420418 -0.33942806]

Multivariate Coefficients: [ 0.62934965 -0.63196673  0.6440201 ]
```

RESULT:	
This step-by-step process will help us to implement univariate, bivariate, and multivariate regression models using the Iris dataset and analyze their performance.	
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